# Filecules and Small Worlds in the DZero Workload: Characteristics and Significance

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Grid: Resource-Sharing Environment

#### Users:

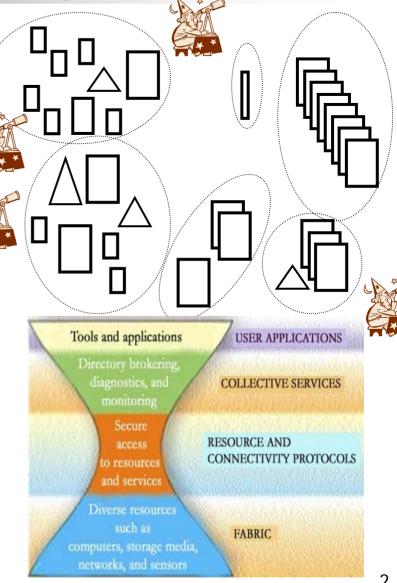
- 1000s from 10s institutions
- Well-established communities

#### Resources:

Computers, data, instruments, storage, applications

Owned/administered by institutions

- Applications: data- and computeintensive processing
- Approach: common infrastructure





#### The Problem

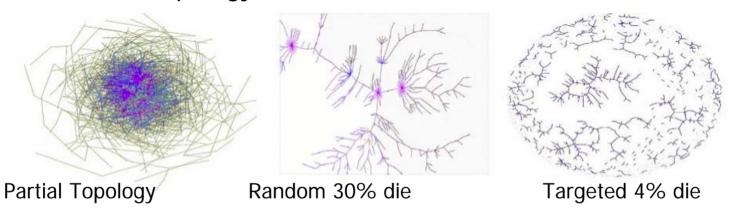
- We have now:
  - Mature grid deployments running in production mode
- We do not have yet:
  - Quantitative characterization of real workloads.
    - How many files, how much input data per process, etc.
  - And thus, benchmarks, workload models, reproducible results
  - Costs:
    - Local solutions, often replicating work
    - "Temporary" solutions that become permanent
    - Far from optimal solutions
    - Impossible to compare alternatives on relevant workloads

## Still, Why Should We Care?

- Impossibility results, high costs: Tradeoffs are necessary
  - Solution: Select tradeoffs based on
    - User requirements (of course)
    - Usage patterns
- Patterns exist and can be exploited. Examples:
- 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 |

# of References

- Zipf distribution for request popularity (web caching) Breslau et al., Infocom'99
- Network topology:



from Saroiu et al., MMCN 2002

# This Presentation

- ...characterizes workloads from DZero from the perspective of data management
  - Data is the main resource shared in many grids
  - High-energy physics domain
  - Potentially representative for other domains
- ...proposes a data abstraction (filecule) relevant to multi-file data processing
- ...identifies a novel pattern (small-world file sharing) relevant to data sharing
- ...shows benefits via experiments
- and invites your comments and suggestions.

## The DØ Experiment

- High-energy physics data grid
- 72 institutions, 18 countries, 500+ physicists
- Detector Data
  - 1,000,000 Channels
  - Event rate ~50 Hz
  - So far, 1.9 PB of data (Update?)
- Data Processing
  - Signals: physics events
  - Events about 250 KB, stored in files of ~1GB
  - Every bit of raw data is accessed for processing/filtering
  - Past year overall: 0.6 PB (Update?)
- DØ:
  - ... processes PBs/year
  - ... processes 10s TB/day
  - ... uses 25% 50% remote computing



## DØ Workload Characterization

Joint work with
Shyamala Doraimani (USF) and
Gabriele Garzoglio (FNAL)

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### DØ Traces (thanks to Ruth and Gabriele)

- Traces from January 2003 to May 2005 1200
- 234,000 jobs, 561 users, 34 domains,
   1.13 million files accessed
- 108 input files per job on average
- Detailed data access information about half of these jobs (113,062)

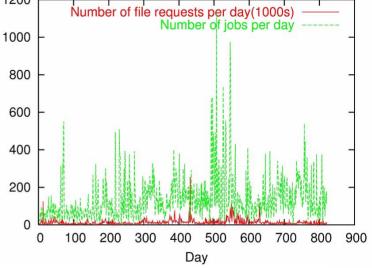


Table 1. Characteristics of traces analyzed per data tier.

Data Tier	Users	Jobs	Files	Input/Job (MB)	Time/Job (hours)
Reconstructed	320	17898	515677	36371	11.01
Root-tuple	63	1307	60719	83041	13.68
Thumbnail	449	94625	428610	53619	4.89
Others	435	120962	N/A	N/A	7.68
All	561	233792	N/A	N/A	6.87



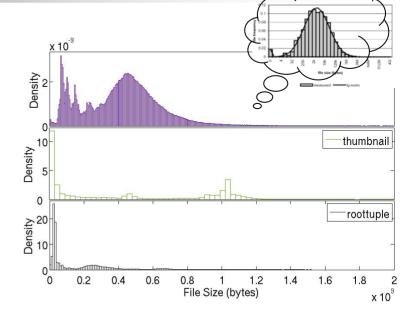
#### **Contradicts Traditional Models**

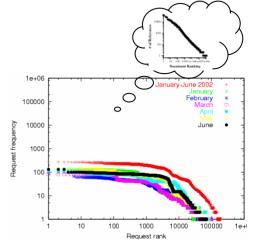
#### File size distribution

- Expected: log-normal. Why not?
  - Deployment decisions
  - Domain specific
  - Data transformation

#### File popularity distribution

- Expected: Zipf. Why not? (speculations):
- Scientific data is uniformly interesting
- User community is relatively small



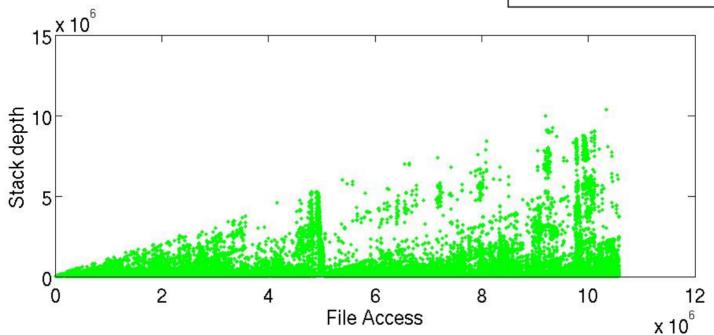


## Time Locality

#### Stack-depth analysis

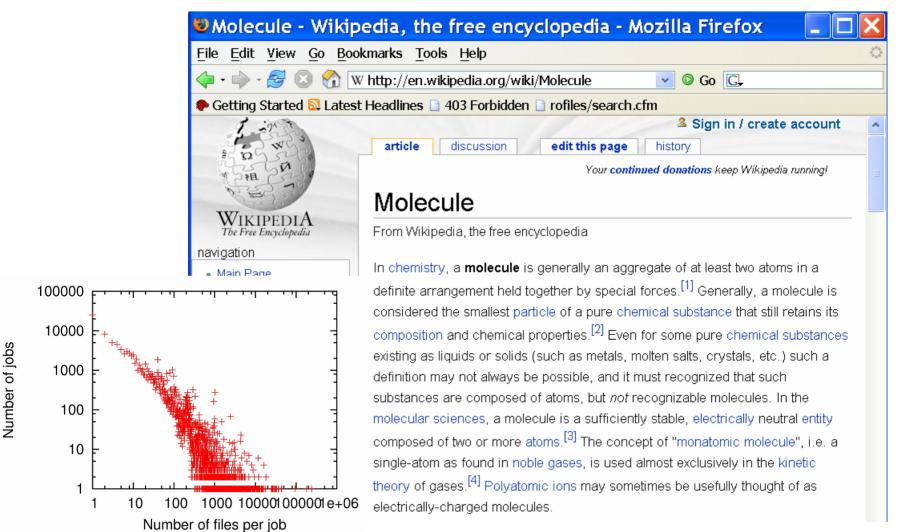
- Good temporal locality
- (to be used in cache replacement algorithms)

Measure	Value		
Maximum	946,600		
1 percentile	85		
10 percentile	960		
50 percentile (Median)	12,260		
90 percentile	90,444		
Standard Deviation	79,300		



## Filecules: Intuition





### Filecules: General Characteristics

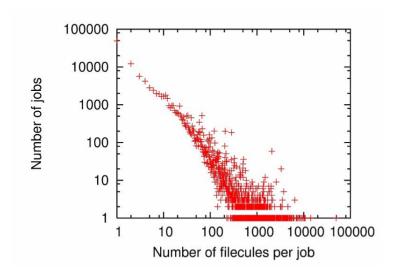
Table 2. Characteristics of analyzed traces per location.

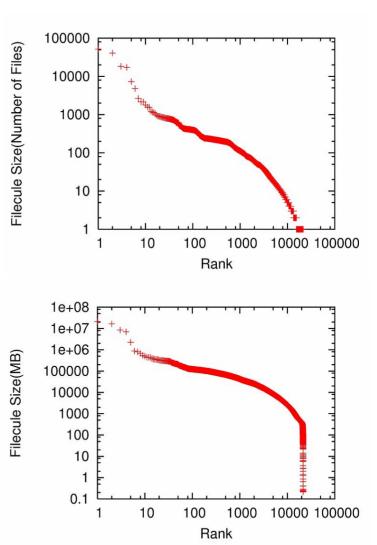
Domain	Jobs	Submission nodes	Sites	# users	# filecules	# files	Total data (GB)
.gov	3319711	12	1	466	95234	945031	4930850
.de	390186	5	4	23	33403	100257	268815
.uk	131760	8	4	21	23876	62427	117097
.edu	54672	18	12	32	14504	36868	41081
.cz	7400	1	1	1	4789	7660	9869
.ca	5719	5	2	4	649	8937	22341
.fr	5086	2	1	11	1767	18215	23958
.nl	3854	3	2	8	888	38812	44012
.mx	146	1	1	1	32	1589	349
.br	12	2	2	2	2	2	2
.cn	4	1	1	2	2	62	31
.in	3	1	1	2	2	2	0.70

## Filecules: Size

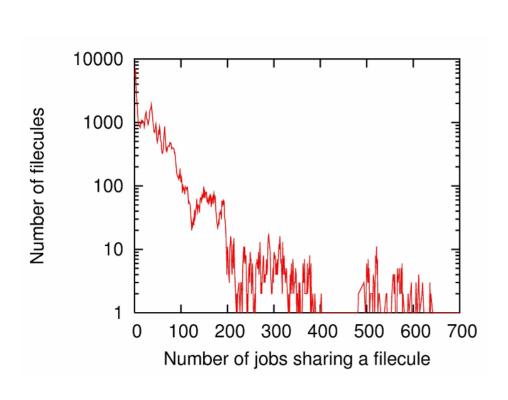
#### Filecules of different sizes:

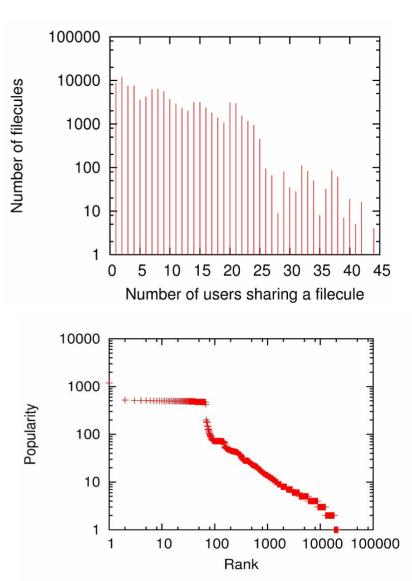
- Largest filecule: 17 TB or 51,841 files
- 28% mono-file filecules





## Filecules: Popularity

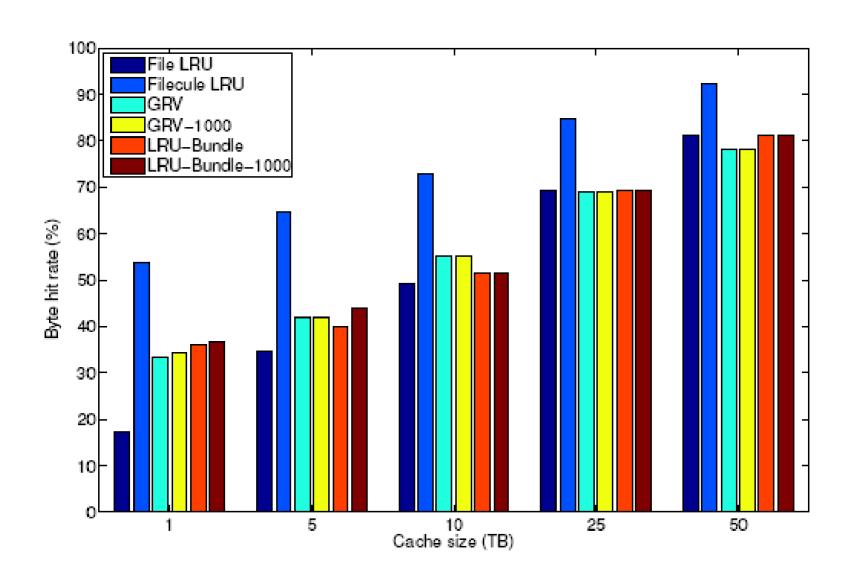




## Consequences for Caching

- Use filecule membership for prefetching
  - When a file is missing from the local cache, prefetch the entire filecule
- Use time locality in cache replacement
  - Least Recently Used (classic algorithm)
- Implemented:
  - LRU with files and LRU with filecules
  - Greedy Request Value: prefetching + job reordering
    - Does not exploit temporal locality
    - Prefetching based on cache content
  - Our variant of LRU with filecules and job reordering
- E. Otoo, et al. Optimal file-bundle caching algorithms for data-grids. In SC '04

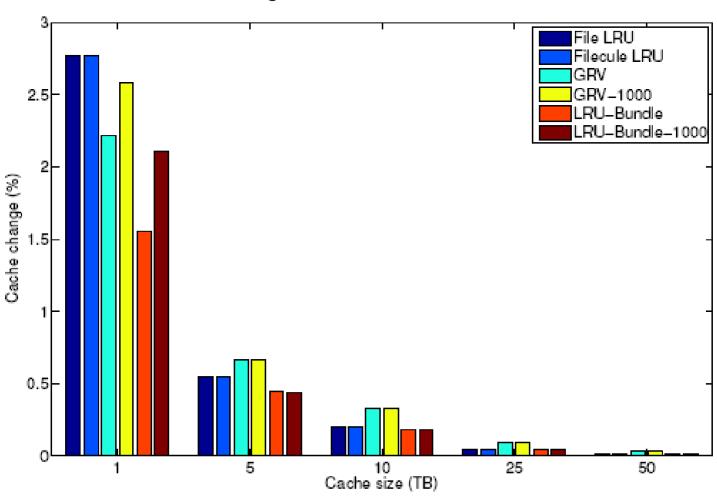
## Comparison: Caching Algorithms (1)





## Comparison: Caching Algorithms (2)

% of cache change is a measure of transfer costs.



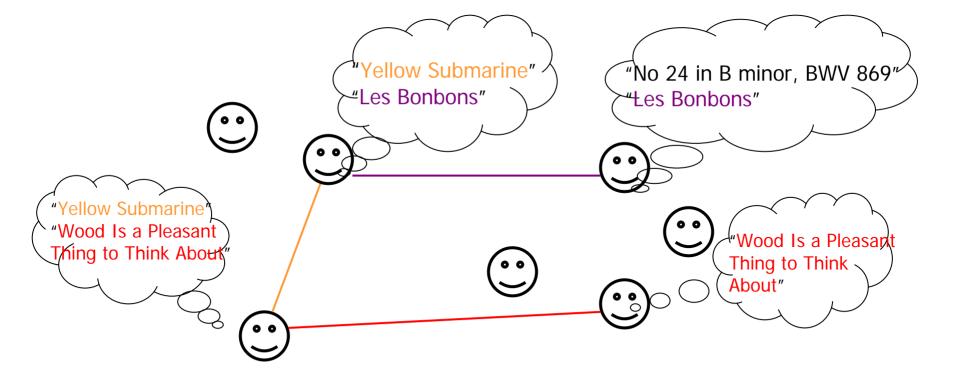
## Summary Part 1

- Revisited traditional workload models
  - Generalized from file systems, the web, etc.
  - Some confirmed (temporal locality), some infirmed (file size distribution and popularity)
- Compared caching algorithms on D0 data:
  - Temporal locality is relevant
  - Filecules guide prefetching

Metric	Algorithm with the best performance
Byte hit rate	Filecule LRU
Percentage of cache change	LRU-Bundle
Job Waiting Time	GRV
Queue Length	GRV
Scheduling Overhead	File LRU and Filecule LRU



Joint work with Matei Ripeanu (UBC) and Ian Foster (ANL and UChicago)

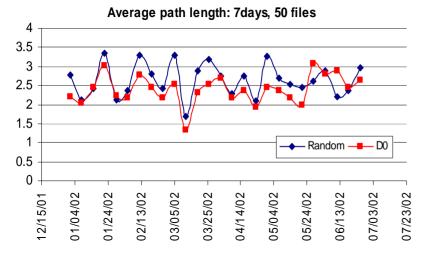


New metric: The Data-Sharing Graph  $G_m^T(V, E)$ :

- $\Box$  V is set of users active during interval T
- An edge in E connects users that asked for at least m common files within T

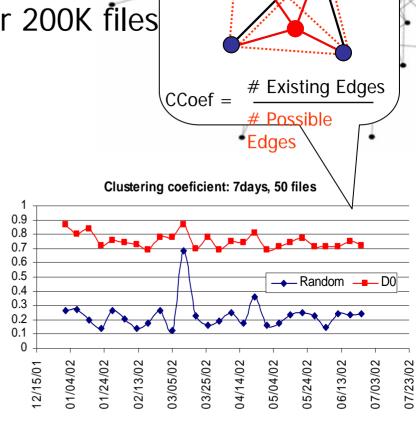
## The DØ Collaboration

6 months of traces (January – June 2002) 300+ users, 2 million requests for 200K files



Small average path length



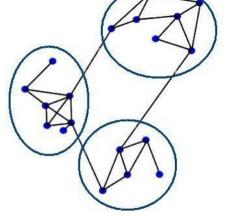


Large clustering coefficient

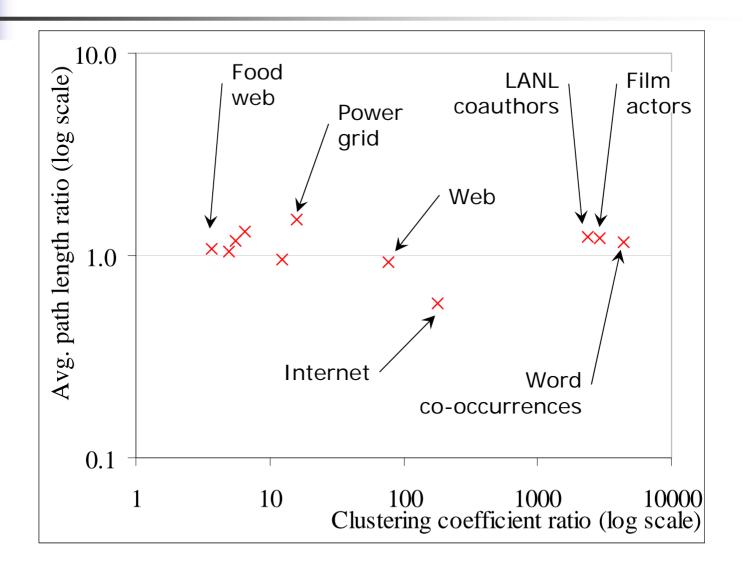


## Small-World Graphs

- Small path length, large clustering coefficient
  - Typically compared against random graphs
- Think of:
  - "It's a small world!"
  - "Six degrees of separation"
- Milgram's experiments in the 60s
- Guare's play "Six Degrees of Separation"



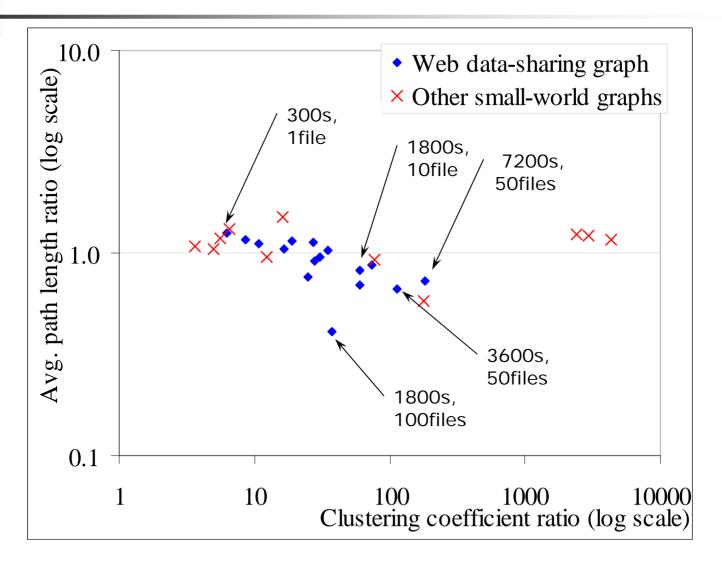
#### Other Small Worlds



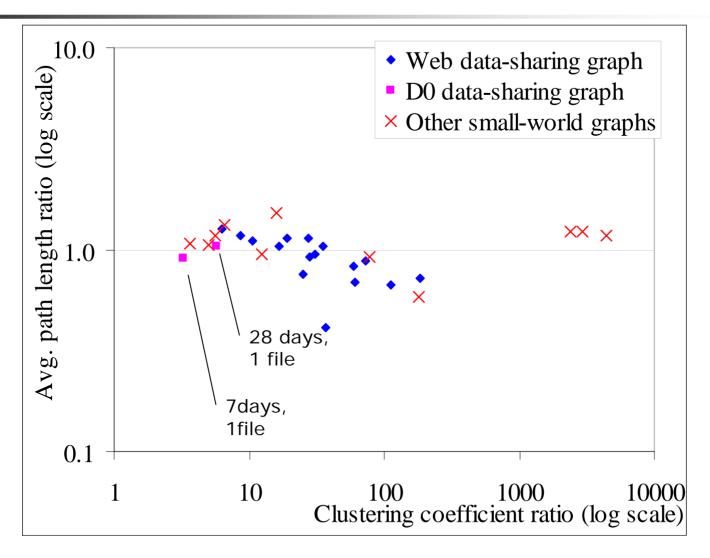
D. J. Watts and S. H. Strogatz, *Collective dynamics of small-world networks*. Nature, 393:440-442, 1998 R. Albert and A.-L. Barabási, *Statistical mechanics of complex networks*, R. Modern Physics 74, 47 (2002).



## Web Data-Sharing Graphs

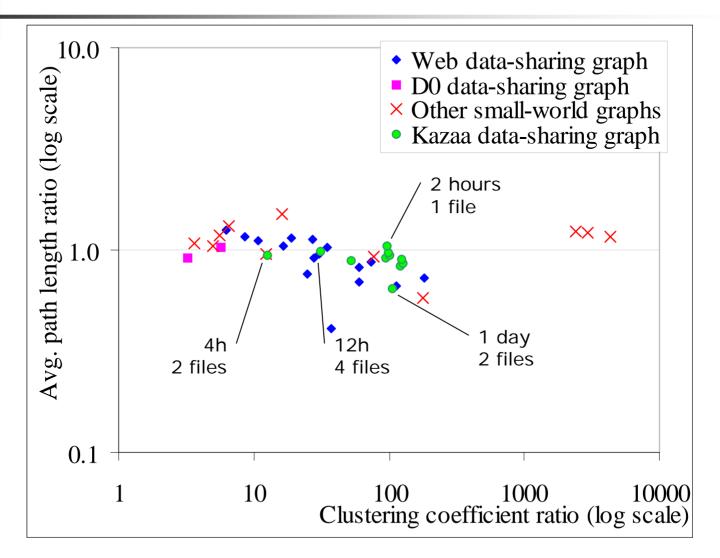


## DØ Data-Sharing Graphs

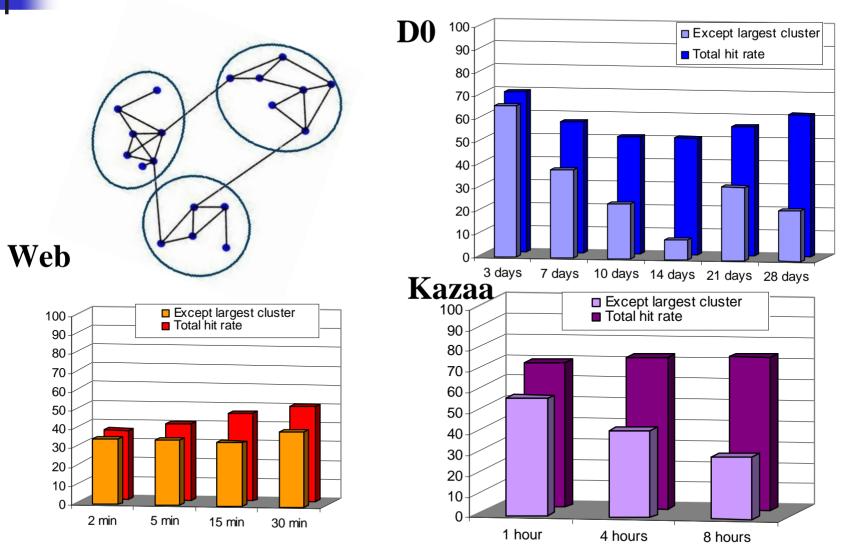




## KaZaA Data-Sharing Graphs



### Interest-Aware Data Dissemination



Interest-Aware Information Dissemination in Small-World Communities, lamnitchi and Foster, HPDC'05

# Amazon's Simple Storage Service: Cost Evaluation for D0

Work with Mayur Palankar, Ayodele Onibokun (USF) and Matei Ripeanu (UBC)



## Amazon's Simple Storage Service

- Novel storage 'utility':
  - Direct access to storage
- Self-defined performance targets:
  - Scalable, infinite data durability, 99.99% availability, fast data access
- Pay-as-you go pricing:
  - \$0.15/month/GB stored and \$0.20/GB transferred
  - Recently updated pricing scheme

Is offloading data storage from an in-house mass-storage system to S3 feasible and cost-effective for scientists?

### Amazon S3 Architecture

#### Two level namespace

- Buckets (think directories)
  - Unique names
  - Two goals: data organization and charging
- Data objects
  - Opaque object (max 5GB)
  - Metadata (attribute-value, up to 4K)

#### Functionality

- Simple put/get functionality
- Limited search functionality
- Objects are immutable, cannot be renamed

#### Data access protocols

- SOAP
- REST
- BitTorrent

# 1

## S3 Architecture (...cont)

#### Security

- Identities
  - Assigned by S3 when initial contract is 'signed'
- Authentication
  - Public/private key scheme
  - But private key is generated by Amazon!
- Access control
  - Access control lists (limited to 100 principals)
  - ACL attributes
    - FullControl,
    - Read & Write (for buckets only for writes)
    - ReadACL & WriteACL (for buckets or objects)
- Auditing (pseudo)
  - S3 can provide a log record



#### Approach

- Characterize S3
  - Does it live up to its own expectations?
- Estimate the performance and cost of a representative scientific application (DZero) in this context
- Is the functionality provided adequate?

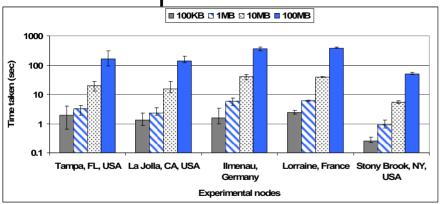
#### S3 characterization methodology

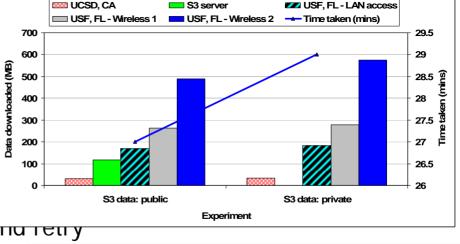
- Black-box approach using PlanetLab nodes to estimate:
  - durability,
  - availability,
  - access performance,
  - the effect of BitTorrent on cost savings
- Isolate local failures

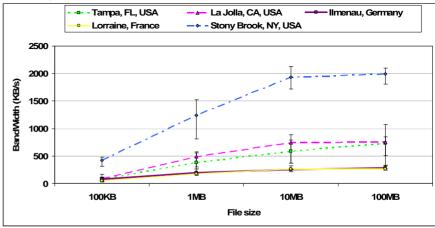
### S3 Evaluation

- Durability
  - Perfect (but based on limited scale experiment)
- Availability
  - Four weeks of traces, about PlanetLab nodes
  - Retry protocol, exponential
  - 'Cleaned' data
    - 99.03% availability after or
    - 99.55% availability after fir
    - 100% availability after second reπy









## S3 Evaluation: Security

#### Risks

- Traditional risks with distributed storage are still a concern:
  - Permanent data loss,
  - Temporary data unavailability (DoS),
  - Loss of confidentiality
  - Malicious or erroneous data modifications
- New risk: direct monetary loss
  - Magnified as there is no built-in solution to limit loss
- Security scheme's big advantage: it's simple
- ... but has limitations
  - Access control
    - Hard to use ACLs in large systems needs at least groups
    - ACLs limited to 100 principals
  - No support for fine grained delegation
  - Implicit trust between users and the service S3
    - No 'receipts'
    - No support for un-repudiabiliy
  - No tools to limit risk

### S3 Evaluation: Cost

- Hypothetical scenario:
  - S3 used by a scientific community: The DZero Experiment
    - 375 TB data, 5.2 PB processed
- Costs
  - Scenario 1: All data stored at S3 and processed by DZero
    - Storage \$675,000/year for storage (\$.15/GB)
    - Transfer \$462,222/year for transfer (\$.20/GB. Now \$.13-\$.18/GB)
      - ⇒ \$94,768 per month!
  - Scenario 2: Reducing transfer costs
    - Caching: With a 50TB cooperative cache → \$66,329 per year in transfer costs
    - Using EC2 → No transfer costs but about 45K in compute costs.
  - Scenario 3: Reducing storage costs
    - Useful characteristic: data gets 'cold'
      - Throw away derived data
      - Archive old data better with S3 support

# Summary

- Workload characterization based on a HEP grid
  - Quantify scale (data processed, number of files)
  - Contradict traditional models
- Patterns can guide resource management
  - Filecules: caching, data replication
  - Small world data sharing: adaptive information dissemination, replica placement

## Thank you.



### Questions

- Storage costs for D0: how do they compared with S3 costs?
- Would you use a storage utility?
- What would you request from a storage utility provider:
  - Usage records: need to be private?
  - Benefits
- Other traces?



